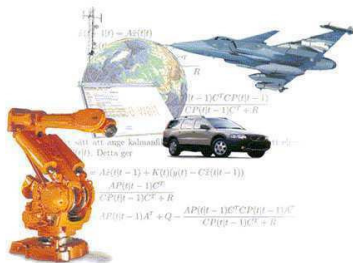


Pedestrian Group Tracking Using the GM-PHD Filter



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- Objectives:
 1. Track the groups of pedestrians
 2. Represent the shape of the groups
 3. Estimate the number of people in each group



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 1. Track the groups of pedestrians
 2. Represent the shape of the groups
 3. Estimate the number of people in each group
- Tested on publicly available video data:
“PETS 2012 dataset S1: Person count and density estimation”.



Frame 50



Frame 100



Frame 150



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1. Group tracking without individual tracks.
 - Treat the group as a single object, do not track single individuals.
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Proposed approach to group tracking;

- I In each image, detect pedestrians
- II Transform image detections to ground positions
- III Use GM-PHD filter to track the pedestrians
- IV Cluster the GM-PHD filter output to form groups
- V Threshold PHD-intensity for each group to represent shape

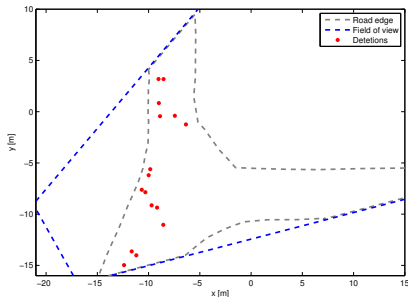
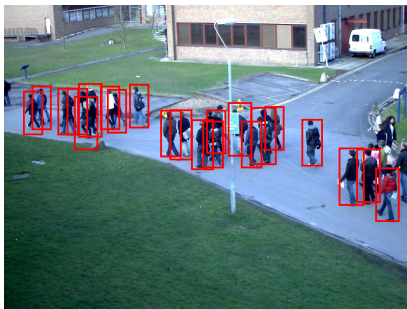




- Publicly available detector by Dollár *et al*, refs in paper
- Low measurement noise
- Missed detections due to occlusions. $p_D = 0.70$
- Few false detections (< 1 per image), i.e. low clutter



- Center point of bottom edge projected onto ground plane.
- Requires:
 1. Calibrated camera – available online with images
 2. Known camera pose – available online with images
 3. Known ground elevation map – here world is assumed flat



- Gaussian mixture PHD filter

$$v_{k|k}(x) = \sum_{i=1}^{J_{k|k}} w_{k|k}^{(i)} \mathcal{N}(x; m_{k|k}^{(i)}, P_{k|k}^{(i)})$$

- State vector is position and velocity: $x = [p^T \ v^T]^T$



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- State vector is position and velocity: $x = [p^T v^T]^T$
- Constant velocity motion model
- Detections are linear measurements of position p_k
- New targets appear where FoV intersect the road



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$$\int_{A,V} v_{k|k}(x) dx$$



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- Pedestrian estimates:

$$\hat{x}_{k|k}^{(i)} = m_{k|k}^{(i)} \quad i : w_{k|k}^{(i)} > \tau$$

- Standard choice $\tau = 0.5$.



- For the extracted pedestrian estimates $\hat{x}_{k|k}^{(i)}$
- Two pedestrian estimates i and j are connected if

1. $\left\| \hat{\mathbf{p}}_{k|k}^{(i)} - \hat{\mathbf{p}}_{k|k}^{(j)} \right\|_2 < 2 \text{ m}$ and
2. $\left\| \hat{\mathbf{v}}_{k|k}^{(i)} - \hat{\mathbf{v}}_{k|k}^{(j)} \right\|_2 < 1 \text{ m/s}$



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- Two pedestrian estimates k and ℓ belong to the same group g if
 1. there is a set of $N \geq 2$ components indexed i_1, i_2, \dots, i_N ,
 2. where $i_1 = k$ and $i_N = \ell$,
 3. and where i_s and i_{s+1} are connected for $s = 1, \dots, N - 1$.

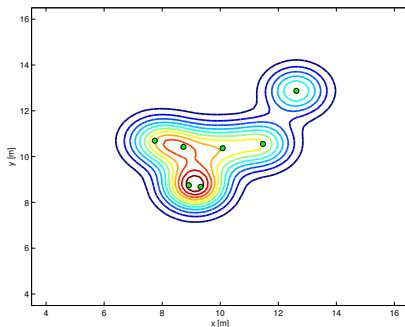
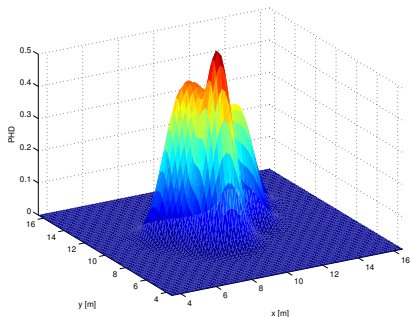


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- The number of pedestrians in a group g is estimated as

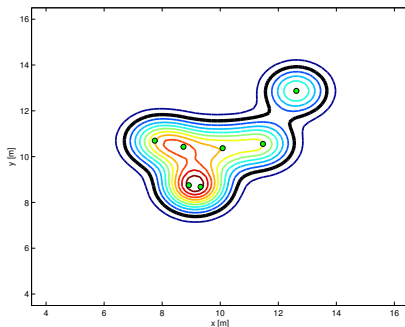
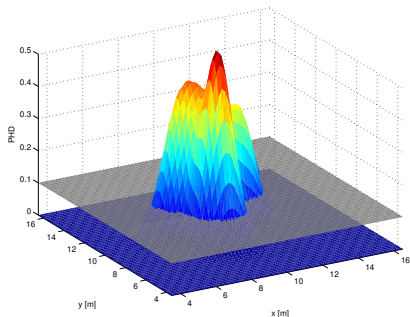
$$\hat{N}_{k|k}^g = \sum_{j=1}^{J_{k|k}^g} w_{k|k}^{g,(j)}.$$



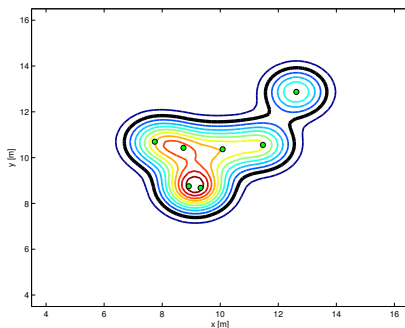
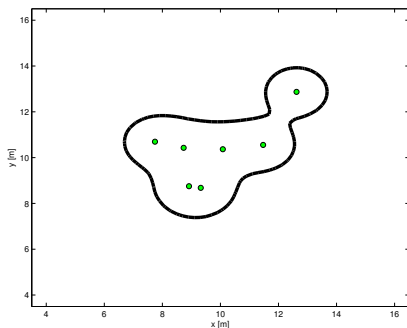
- For group g the PHD-intensity is $\sum_{i=1}^{J_{k|k}^g} w_{k|k}^{g,(i)} \mathcal{N} \left(x; m_{k|k}^{g,(i)}, P_{k|k}^{g,(i)} \right)$



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- PHD is intersected at 0.1
Intersection is approximation of group's size and shape.

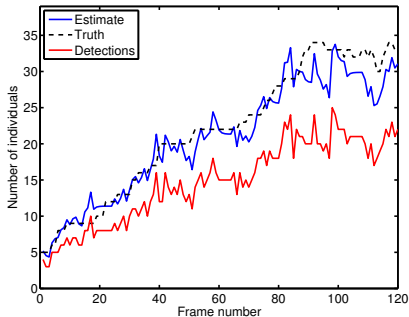


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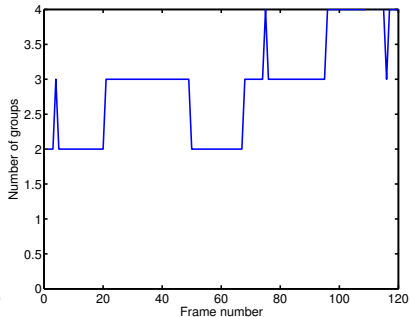


<http://youtu.be/aAz3poW49CU>





Number of pedestrians



Number of groups



- Pedestrian detector used on each image
- Detections projected onto ground plane
- Projections input in GM-PHD filter to get pedestrian estimates
- Groups formed using clustering
- Group shape represented by thresholded PHD-intensity



Recent work

- Gaussian mean shift clustering to form groups
- Group IDs for continuity
- See Viktor Edman's Master's Thesis

<http://urn.kb.se/resolve?urn=urn:nbn:se:liu:diva-93996>

<http://youtu.be/7DEkaTeFLWY>

Future work

- Compute uncertainty of image-to-ground plane projection
- Compare results with extended target PHD-filter
- Include also group dynamics, not just individual dynamics
- Dynamic way of forming groups



Thank you for listening!

Any questions?

